EMOTIONAL MACHINE ORCHESTRATOR

By Point of View

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THE TEAM



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THE SMALL PROBLEM

- Many text generation platforms are apathetic or unpredictably emotional. It would be useful to be able to generate specific emotional text.
- This could be useful for chatbots presenting a specific emotion or for realistic emotional dialogue creation.

THE BIG PROBLEM

- A major goal of Al Practitioners is to create a system that can generate creative fiction.
- A subset of this work would be the creation of a dialogue system that builds on specific character archetypes and situations.
- Emotional modeling could be used in conjunction with personality modeling, dialect modeling, etc. to build distinctive characters and to create a story trajectory that is meaningful.

OUR SOLUTION

- We built a package of six different emotional text generation models, collectively referred to as EMO: Emotional Machine Operator.
- Each model is associated with a specific emotion





USING THE MODEL

Option 1: An emotion is designated



A "Call" is entered as an input



Multiple options for "Responses" are displayed

Option 2: An arc of emotions is designated



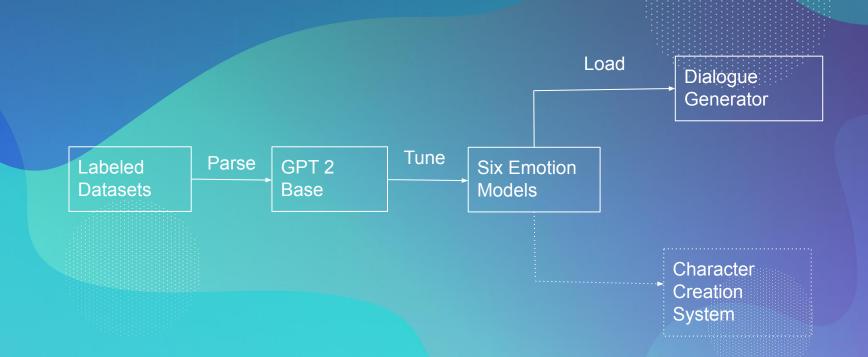
A "Call" is entered as an input



A single set of dialogue following the emotional arc is displayed

DEMO: Tackling the Small Problem

PIPELINE



DIVING DEEP

- Preliminary research to determine which emotions to include
- Models are built on GPT-2, a tunable open source language model from OpenAl
- Combined two datasets to finetune our models
- Optimized our tuning using quantitative data results.
- Developed a Python script that accepted a set of emotions and loaded our EMO models to create a dialogue.

CHALLENGES

- Navigating GPT-2
- Time constraints of fine-tuning GPT-2
- Sourcing sufficiently large, detailed datasets (right)
- Optimizing GPT-2-simple's fine tuning parameters
- Increasing context-guided content of responses

DailyDialog: A Manually Labelled Multi-turn Dialogue Dataset

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log dataset, DailyDialog, which is intriguing in several aspects. The language is human-written and loss noisy. The dialogues in the dataset reflect our daily com-

1 Introdu demic camps enterprises to and thus redu it is challeng an intelligen high-level na

area are still tems are ofter dialogue data OpenSubtitle

A: Well, I have to drive to school for a meetin this morning, and I'm going to end up getting stuck in rush-hour traffic B: That's annoving, but nothing to worry about Aust breathe deeply when you feel yourself getting

EmotionLines: An Emotion Corpus of Multi-Party Conversations

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1. Introduction

oriented dialogue system and the social (chit-chat) dialogue system. The former focuses on designing a personal assistast which can accomplish certain tasks, and for the la-ter it is important to capture the conversation flow which emphasizes more on the feelings of the speaker. Many re-searchers my to baild a "neura" dialogue system by enhan-ing dialogue breadth (coverage), dialogue depth (comples-ity) or both. Those who want to increase dialogue beauth try to transfer dialogue acts across domains (Chen et al., 2016) to establish multi-domain or even open domain dia-2016) to establish radii-denais or ewn open domain dis-logue system, and those who want to deepen dislogue com-plexity pay their attention to transform a knowledge-based systems to common some or even empathetic systems that can recognize contion features, generate enaction-aware responsor (Fung et al., 2016), or learn how to plan the dislogues while users interact via high-level descriptions (Sun et al., 2016). No matter what kind of dialogue system we want to build, a useful and large dialogue dataset is indis-

pensable.

When building a task-oriented dialogue system, dialogue When building a task-oriented dialogue system, dialogue corpora with dialogue act information is accessible and hence are commonly utilized. However when building a chit-chat conversational bot, though the importance of emotion detection has been noticed, pure conversation con-tent such as movie, TV scripes or chat logs without em-tion labels are more available: no emotion labels on ut-terances can be used for learning. Moreover, when we turn to other datasets with amorated emotion information such as data crawled from Twitter [Mcharmand and Brave-lations.] endent. As a result, models built with these datasets lack the ability to consider contentual information essential

These authors contributed equally to this work.

in dialogue systems, not to mention the ability to captur

Just got back from seeing @GaryDelaney in Burslem, AMAZING!! Face still burts from Joy laughing so much #hilarious English sangthless as alsony differentiate I get so nervous even thinking about talking to *** *** I wanna die Work use using @mothercareak @Methercarehelp again!! These guys cant Augur get nothing right!! #furning

Table 1: Emotion labeled posts without contextual informs tion (selected from WASSA-2017 Shared Task on Emotion Intensity)

textual information may encounter another issue that the

same utterance can express catterest emotions depending on its context. Table 2 shows some examples of saying "Okay!" with different emotions. The IEMOCAP database (Busso et al., 2008), to the best of our knowledge, is the only dataset that provides emotion labels for each atterance. However, IEMOCAP was created by actors performing emotions, and hence carries the risk of overacting. Moreover, the annotators tabel the emotions by watching the videos instead of reading the transcripts which nears the annotators may make the decision only depend on the facial expression or the prosodic features without

To tackle these problems, we create EmotionLines: as

FUTURE WORK

- Fine tuning GPT-2 on larger data sets and/or using automated sentiment analysis to generate larger, more diverse training files
- Evaluating accuracy of generations with automated sentiment analysis
- Increasing tolerance of multi-line input
- Using fine tuned models to create a framework for building full scripts
- Engaging in more ethical analysis of potential (mis)use cases
- Lots more!

FURTHER READING / SOURCES

- OpenAl's GPT-2
- Mark Woolf's <u>gpt-2-simple</u> setup
- Datasets: <u>EmotionLines</u> and <u>DailyDialogue</u>
- YouTuber <u>Tom Scott</u> uses GPT-3 to generate video ideas
- Readings on ethical considerations:
 - On the Dangers of Stochastic Parrots: Can Language Models Be Too Big?
 - Training a single Al model daniemit as much carbon as five cars in their lifetimes.

Thank you!

Question? Comments? Suggestions?

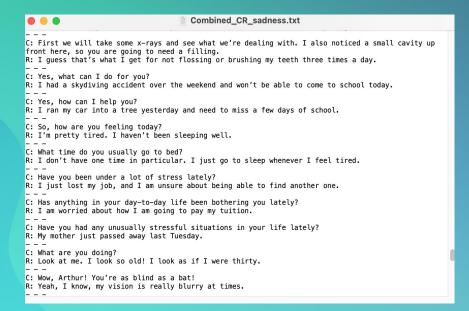
BONUS SLIDES

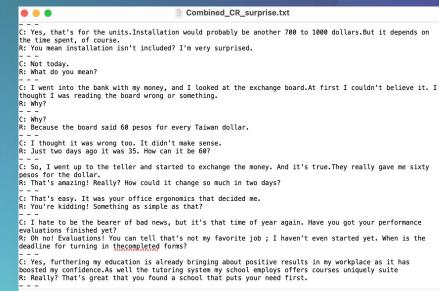
VALIDATION: Training Data & Joyful Model

- At *temperature* = 0.9, in a batch of n = 20 generations of the joyful model:
 - about 64% of generated utterances are not identical to utterances in the training data
 - o about 95% of generated utterances are not identical to other utterances in the batch

	А	В	С	D	Е	F	G
1	Fintuned Model	steps (fintuning)	nsamples	temperature	prefix	SELF - % unique	TRAIN - % unique
2	Combined_CR_happiness					85.70%	(BASELINE)
3	Combined_CR_fear					86.89%	(BASELINE)
4	Combined_CR_sadness					91.27%	(BASELINE)
5	Combined_CR_anger					89.02%	(BASELINE)
6	Combined_CR_disgust					94.12%	(BASELINE)
7	Combined_CR_surprise					87.18%	(BASELINE)
8						89.03%	
9	Combined_CR_happiness: test_run_2	500	20	0.9	New Zealand	100.00%	67.86%
10	Combined_CR_happiness: test_run_3	500	20	0.9	New Zealand	96.30%	70.37%
11	Combined_CR_happiness: test_run_4	500	20	0.9	New Zealand	92.00%	64.00%
12	Combined_CR_happiness: test_run_5	500	20	0.9	First day	93.33%	66.67%
13	Combined_CR_happiness: test_run_6	500	20	0.9	First day	100.00%	53.33%
14						96.33%	64.45%

SAMPLE TRAINING DATA





Given a call as input ("C: ... "), we want each fine tuned model to generate a response ("R: ... ") that is characteristically joyful, sad, fearful, angry, disgusted, or surprised.

EXAMPLE GENERATIONS TO: "C: So, I'm visiting my aunt in New Zealand next week.

R. "

JOY

Oh, good. You're going to see a lot of people from your country, who are very interested in travelling to the world's oceans.

R: You're so lucky

R: It sounds like fun

R: Oh, that's wonderful

R: It's delightful to be in the country.

R: Oh, that's wonderful. I really hope you can make it to visit her in time.

R: That's great. I'm very interested to know what she's up to

R: That's great. Would you like to meet him?

R: Oh, what's that like?

R: Oh. congratulations, congratulations!

R: Don't worry. I'm free next Monda

R: It's a beautiful time to be a New Zealander! I hope you're bringing along a good trip souvenir.

R: Oh, I see. Well, if you'll excuse me, I should be back by then.

R: It sounds like a good trip!

R: How's your trip going?

R: You are so lucky! I hope to see you again.

R: That's terrific!

R: Yes. it's excellent!

R: Oh, that's great. I just searched on the internet.

R: What a coincidence

R: That's perfect. You'd have to come with me.

R: Wow, what a beautiful trip!

R: We should come to a game, too.

R: Oh, I see.

SADNESS

So sorry, I'm out of sorts at the moment

R: It's rainy weather today and I feel a little bit let down by the scenery

R: Hey, chill out, I'm just trying to help.

R: So sorry, I must have been watching the news

R: It's rainy weather today and I have to go to the store to

R: And I find it's the kindest thing to do

R: And I'm 18 months old

: It's just too hard-hitting and depressing

R: and I have to tell you that I loooove it.

R: So, I'm gonna miss you too.

R: Hey, I'm sorry, I haven't seen you for a long time.

R: So sorry, I gotta go break up with Bonnie now.

R: I am sorry, I'm too tired now. I go to sleep when I see the red light.

R: I'm sorry, I'm going to meet some friends there.

R: And I am worried about my visa.

R: So sorry, I must have been looking for you.

R: Hey, I'm sorry, I-I didn't know. Hey, come on.

R: And I feel a little sad because like, thirty yuan or

R: And I'm 18 right now and I have to leave it so I can't return it to you.

R: I'm sorry, I'll be here until 10 again tonight

R: Hey, if you can wait until tomorrow night, I'll go with you.

R: I am sorry, I just. Look, I-I'm gonna be back to my room in as little as a second

R: I'm sorry, I was really looking forward to this.

R: And I wonder if I could borrow the car from here instead of aging to the girport

ANGER

Are you crazy? You're losing out on a great opportunity here...

R: Come on! Why are you so bitter?

R: Wait a moment, mom.

R: Waiter! I can't believe it! I can't believe it! I've been so busy

R: Wait a moment, her name is Nellie? Don't you know

R: Are you crazy? Youandroid?

R: Come on! Why are you so bitter?

R: Egg! I'm not allowed here

R: Wait, who's coming to visit you, then?

R: Where? bawl!

R: Come on! Why are you so bitter?

R: what?!

R: Come on! Why are you so bitter?

R: Come on! Why are you so bitter

R: Egg! I told you

R: Waiter! I don't understand!

R: Egg! I'm totally nuts! I must be attacked by rabid cats!

R: Come on! Why are you so rude to your wife?

: Waiter! I don't understand!

R: Nooo! Nooooo! Noogie! Stop it! It's expired!

R: You have to be kidding me! Go away! Stop having so many funny pictures!

R: Ross, I'm sorry, but I'm not going to be in touch with

R: Hev. honev! Don't saueeze it so hard

t Noo!! I don't care! I'm not, the wires have come loose in Your head!

R: Hey, that's not a good look on your par

R: Nooo! Nooooo! Noogie! Sing

EXAMPLE GENERATIONS TO: "So, I'm visiting my aunt in New Zealand next week."

SURPRISE

Oh my God! So you're gonna stay here with me?

- R: Oh, really?
- R: Oh. I have your aunt's itinerary
- R: Oh, my as Taiwan
- R: Oh my God! How the hell do you do that?
- R: Oh really?
- R: What?! Look how crazy it is to live in a world of this
- R: Oh, my God! What a great way to start a new year!
- P: Ob roally?
- R: Oh really?
- R: Oh my God! I have an uncle in Thailand
- R: You told me in advance you were going to
- R: Oh, really?
- R: Oh, really
- R: Oh really? Are you in Taiwan?
- R: Oh my God! What's she doing here
- R: Oh, my.
- R: Oh really?
- R: Oh, my God! What a horrible guy.
- R: Oh no! What do you get out of this
- R: Oh, really?
- R: Oh, really? I hope you are going with her.
- R: Oh really
- R: You told me in advance you were going to visit, that you were going to visit all over the world
- R: Oh really?
- R: Oh, really?
- R: Oh, my

FEAR

She's such a pain in the neck!

- R: She is really pretty, isn't she? Her skin looks so baby-smooth!
- R: Is it really your turn this week?
- R: You are not very funny!
- R: It's a pleasure to meet you
- R: She's almost medium
- R: It's no use complaining.If we had got enough mone we wouldn't put up with it!
- R: You are driving me up the wal
- R: Is this a restaurant that you want to come back to
- R: What a terrible thing to do
- R: Her husband is a fat cat!
- R: Just like yesterday. She's making me sicl
- R: She lives just off the coast!
- R: She's really pretty, isn't she? Her skin looks so baby-smooth!
- R: She said she was too excited to go to sleep.
- R: She's really pretty, isn't she? Her skin looks so
- R: She's very sick. I'm just coming from the hospital.
- R: She's gone missing
- R: She's very sick. I'm just coming from the hospital
- R: It's a pleasure to meet you. Happy to help you out!
- R: She's such a pain in the neck!
- R: It's a terrible thing to do in my native tongue.
- R: Sorry to trouble you. But I don't think it's right
- R: She's very sick. I'm just coming from the hospital.
- R: She's here with a bunch of pot bellies who all smell bad. Sorry to trouble you. But I don't think this fish is free It actually tastes a bit off.

DISGUST

Oh, how awful!

- R: She's really nice, isn't she? Her smile is killer!
- R: Is it really your turn to be frank
- R: She's gone to get drunk
- R: She said she was too excited to go to sleep
- R: She's here with her husband. And they've got out a control.
- R: You are not very funny. I doubt you will becom friendly with me again long.
- R: She's very sick. I'm sure she'll be soon die
- R: Can you imagine what would be like if we had got int this game?
- R: She's such a pain in the neck!
- R: She's here with three weeks leave
- R: She's really pretty, isn't she? Her skin looks so
- R: She's a terrible gossip.
- R: She's really pretty, isn't she? Her skin looks so baby-smooth!
- R: She's going to be seeing he
- R: It's a pleasure to meet you.Love
- R: What a terrible thing to do
- R: You are not very funny. I can't believe I married c woman like you.
- R: It's a pleasure to meet you.
- R: It's a very bad idea
- R: She's too excited to take a nap.
- R: She is very pleased with m
- R: It's no use complaining.If we met face to face, he would certainly get along with me.